

### 4.3: Modeling Linear Trends

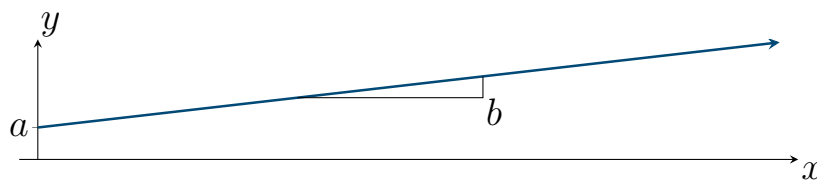
#### Definition.

The **regression line** is a model used for making predictions about *future* observed values. The equation of the regression line is

$$y = a + bx$$

(when  $x=0$ )

where  $a$  is the  **$y$ -intercept** and  $b$  is the **slope**.



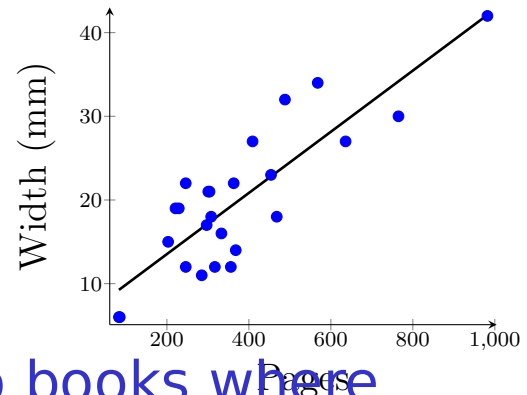
- The input variable  $x$  is also known as the
  - Independent variable
  - Predictor variable
  - Explanatory variable
- The output variable  $y$  is known as the
  - Dependent variable
  - Predicted variable
  - Response variable

**Example.** Below is a scatterplot comparing number of pages a book has against the width of the book. Interpret the intercept and the slope of the regression line.

$$\text{Predicted Width} = 6.22 + 0.0366 \text{ Pages}$$

**Intercept:** 6.22mm  
Width of a book with  
zero pages

**Slope:** 0.0366 mm per page  
When comparing two books where  
one book has an additional page,  
the other is 0.0366mm thicker

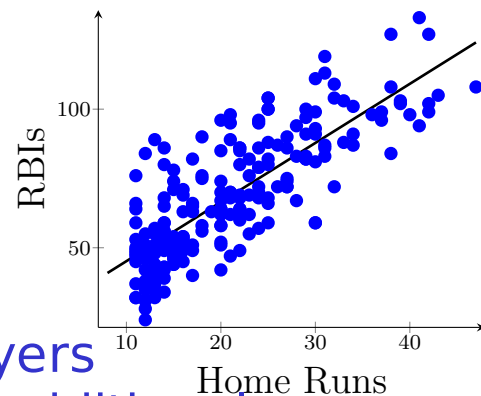


**Example.** Below is a scatterplot comparing the number of home runs and RBIs in the 2016 season. Interpret the intercept and slope of the regression line.

$$\text{Predicted RBI} = 23.84 + 2.13 \text{ HR}$$

**Intercept:** 23.84 RBI  
Could make sense, but is  
outside the data range

**Slope:** 2.13 RBI/HR  
When comparing two players  
where one player hits an additional  
HR, their predicted RBI is expected  
to be 2.13 RBI higher



**Definition.**

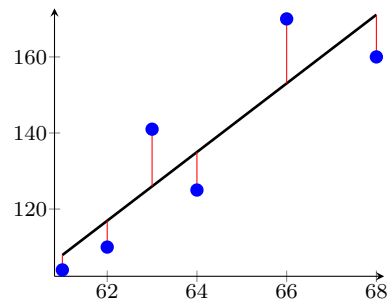
Now we define the formula of the regression line:

$$y = a + bx$$

Where

$$b = r \frac{s_y}{s_x} \quad \text{and} \quad a = \bar{y} - b\bar{x}.$$

These formulae minimize the residual error: [Try this!](#)



**Example.** Below are the heights and weights of six women:

Heights	61	62	63	64	66	68
Weights	104	110	141	125	170	160

From this we get

$$\bar{x} = 64$$

$$s_x = 2.608$$

$$\bar{y} = 135$$

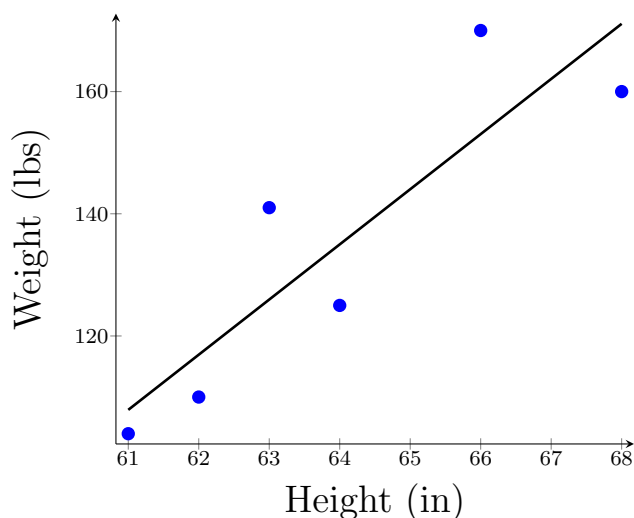
$$s_y = 26.728$$

$$r = 0.881$$

Find the equation of the regression line.

$$b = 0.881 \cdot \left( \frac{26.728}{2.608} \right) = 9.03$$

$$a = 135 - 9.03(64) = -442.92$$



**Example.** Open the `popdensity_and_crime` dataset in StatCrunch, use the “Simple Linear” tool under the `Stat>Regression` menu to find the regression line for the following columns. Interpret the slope and intercept where appropriate.

the `pop1990` and `pop2000` columns,

$$\text{pop2000} = 12266.759 + 1.1295246 \text{ pop1990}$$

R (correlation coefficient) = 0.99649554

Intercept: If `pop`=0 in 1990, it's 12,267 in 2000  
(Not appropriate since data is 454k-30mil)

Slope: When comparing `pop` in 1990 where one group has 1 more person, that group is expected to have 1.13 more people in 2000

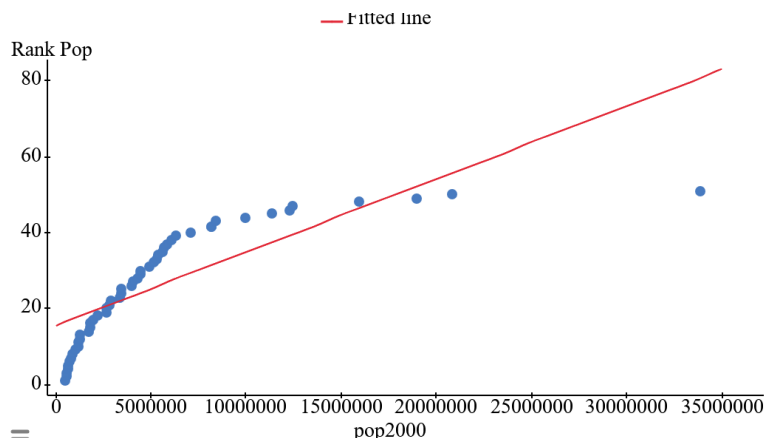
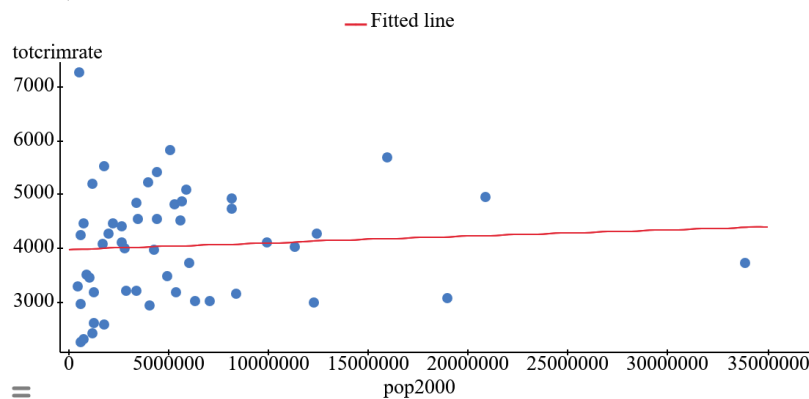
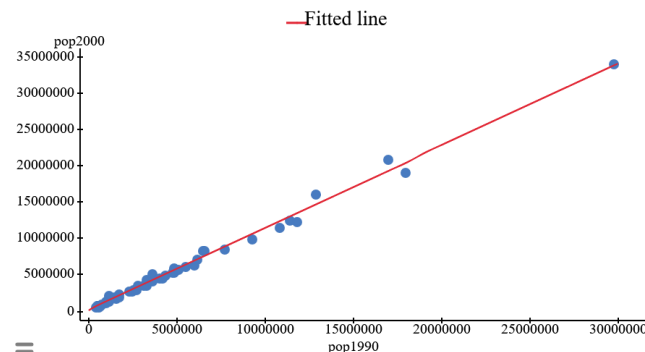
the `pop2000` and `totcrimrate` columns, and

Low correlation coefficient (~0.07)  
Looks nonlinear  
Linear regression is not appropriate here

the `pop2000` and `Rank Pop` columns.

Large correlation coefficient (~0.8),  
but a nonlinear trend

Linear regression is not appropriate



## 4.4: Evaluating the Linear Model

Guidelines:

- Don't fit linear models to nonlinear associations!
- Correlation is not causation
- Beware of outliers (a.k.a. **influential points**)
- Don't extrapolate (make predictions beyond the range of the data)

### **Definition.**

The **coefficient of determination** is the correlation coefficient squared:

$$r^2$$

This is sometimes also called ***r*-squared**.